# Falcon 9 – Data Science Project

Martin R.

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# **Executive Summary**

- The goal of this presentation is to display the prediction of landing of Falcon 9 based on data-driven insights.
- The findings were obtained through:
  - 1. Webscraping data collection
  - 2. Data Wrangling
  - 3. EDA Analysis with SQL
  - 4. Matplotlib and Folium
  - 5. Machine Learning and Predictive Analysis

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## Introduction

#### **About SpaceX**

 SpaceX is one of the most successful spacecraft company owned by Elon Musk. This company has many accomplishments such as sending Starlink internet satellite to the space or sending spacecraft to the International Space Station. One of the reason why SpaceX is so successful is because their rocket launches are not much expensive. That is because their Falcon 9 can reuse the stage.

#### **Goal of this project**

 The goal of this project is to determine the future of rocket launches using data science.

# Methodology

- Data Collection
- Data collection is how we gather data in this project these tools were used:
- API
- Webscraping BeautifulSoup
- Data Wrangling Pandas
- SQL analysis

- Data Preparation
- Data preparation is how we cleanse and prepare data in this project these tools were used:
- Numpy
- Matplotlib
- Seaborn
- Machine Learning

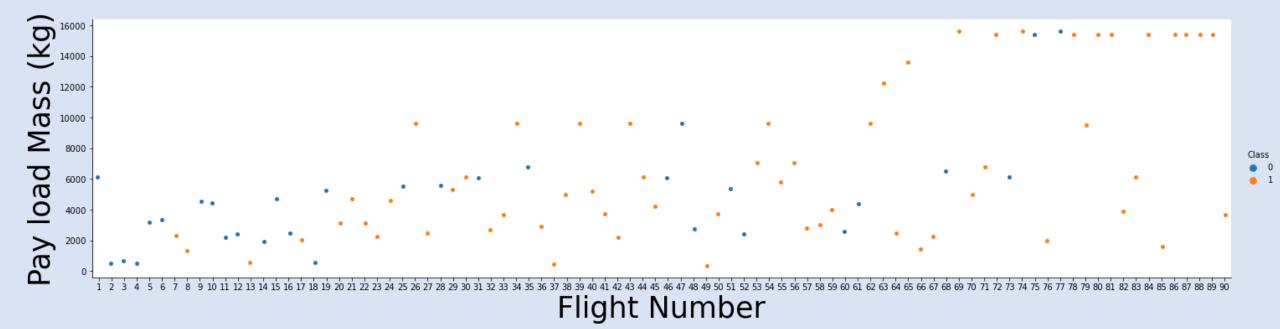
- Data Modeling
- Data modeling is presenting data in this project these tools were used:
- Folium
- Plotly

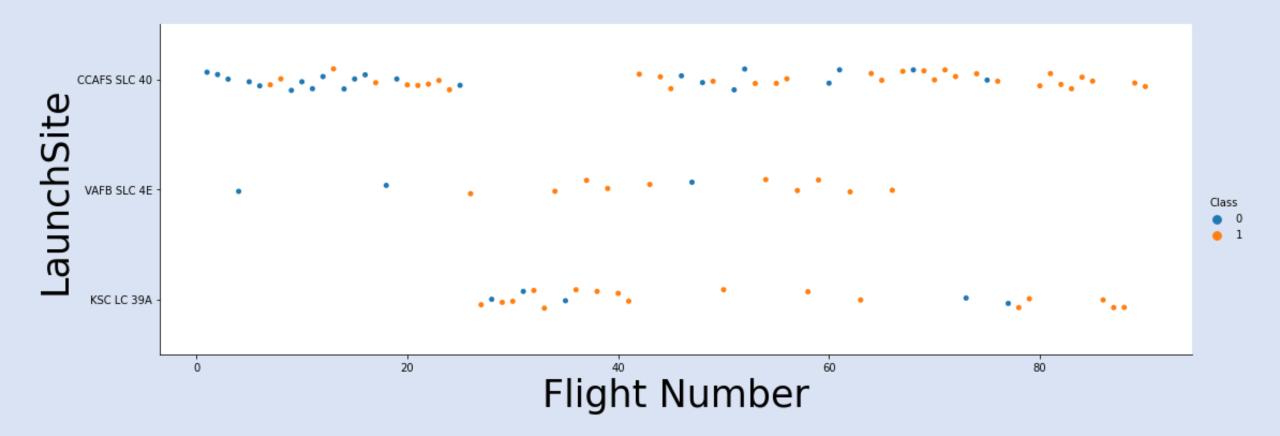


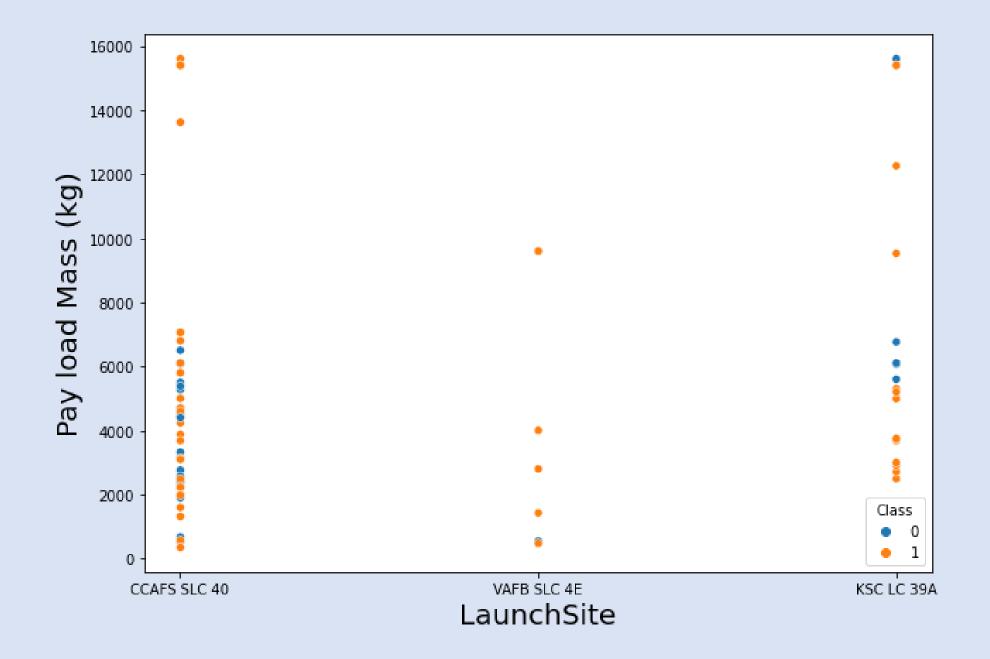
# Webscraping

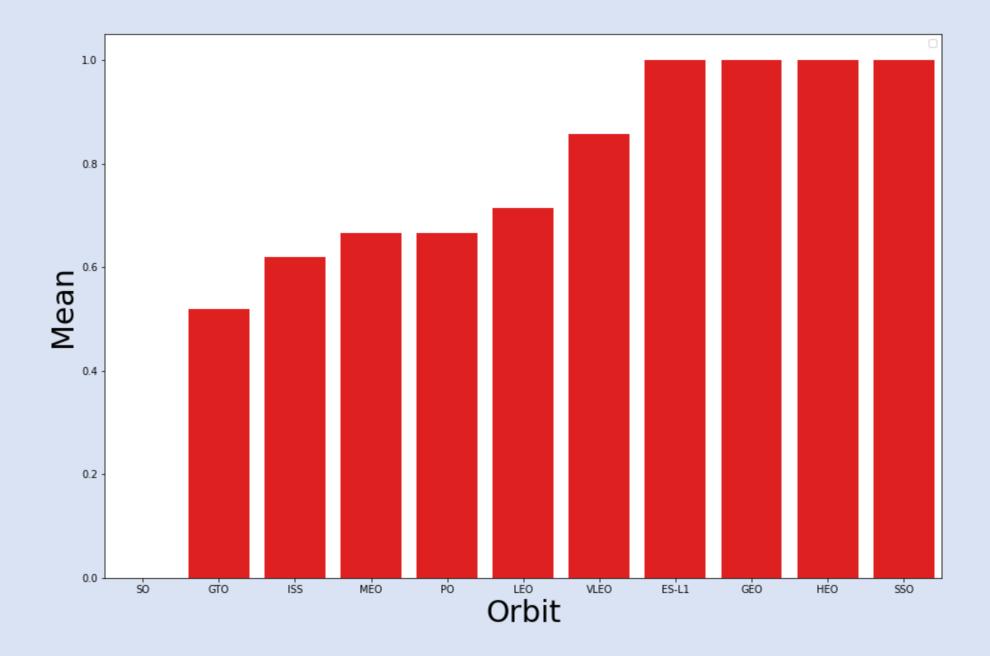
```
launch_dict['Launch outcome'].append(launch_dict)
           print(launch_outcome)
           # Booster landing
           # TODO: Append the launch_outcome into launch_dict with key `Booster landing`
           booster_landing = landing_status(row[8])
           launch_dict ['Booster landing'].append(launch_dict)
           print(booster_landing)
4 June 2010
18:45
F9 v1.0B0003.1
CCAFS
Dragon Spacecraft Qualification Unit
Dragon Spacecraft Qualification Unit
LEO
SpaceX
Success
Failure
8 December 2010
15:43
F9 v1.0B0004.1
CCAFS
Dragon
Dragon
```

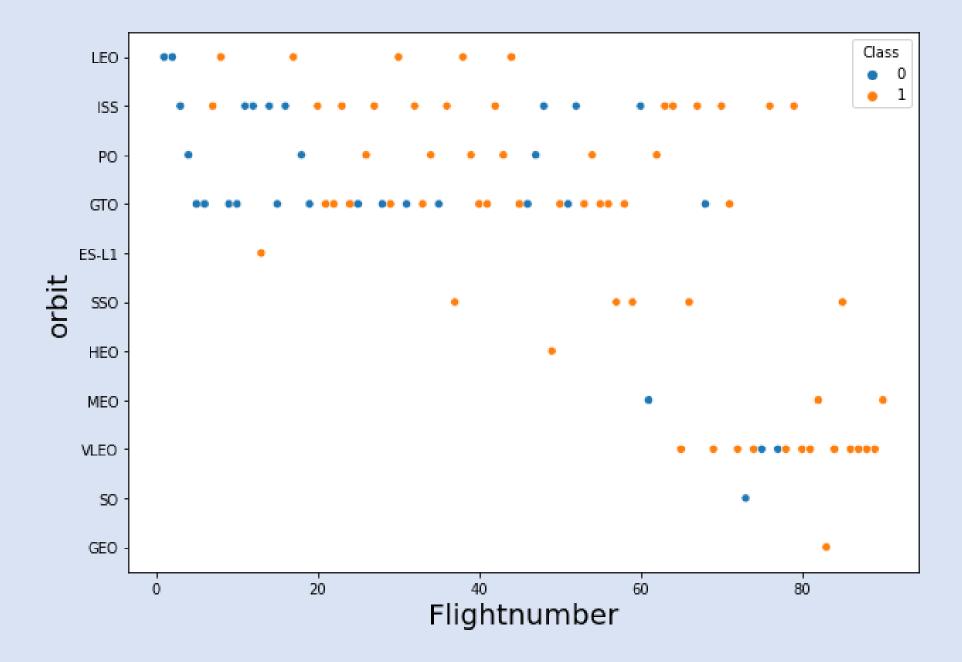
# Exploring and Preparing Data

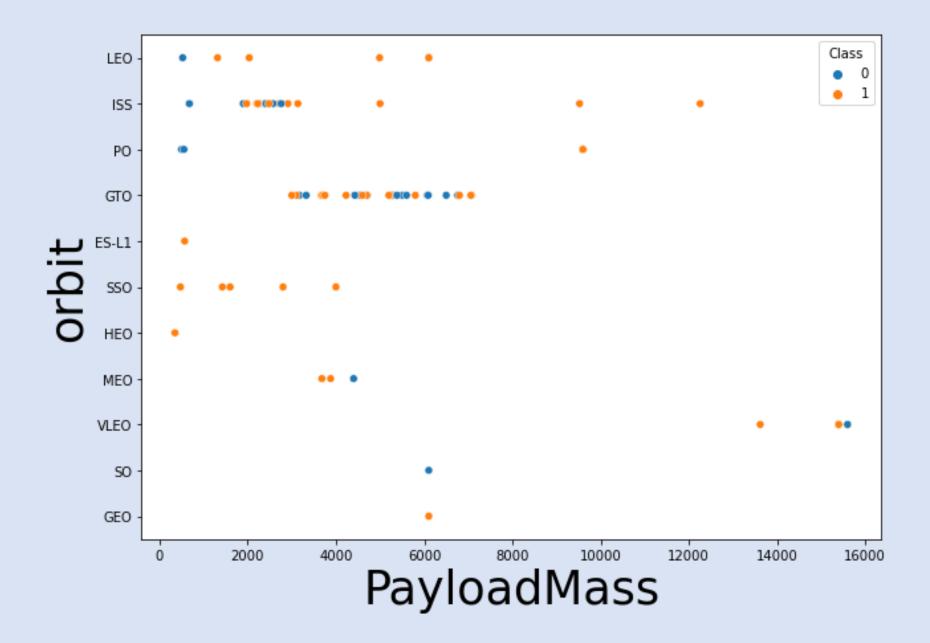












# SQL analysis

#### Display the names of the unique launch sites in the space mission

In [21]: %sql select distinct (launch\_site) from Spacex

 $* ibm\_db\_sa://ybt98819:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludbDone.$ 

Out[21]: launch\_site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

#### Display 5 records where launch sites begin with the string 'CCA'

In [30]: %sql select \* from Spacex where launch\_site like 'CCA%' limit 5

 $* ibm\_db\_sa://ybt98819:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludbDone.$ 

	Done.									
Out[30]:	DATE	timeutc_	booster_version	launch_site	payload	payload_masskg_	orbit	customer	mission_outcome	landingoutcome
	2010- 04-06	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
	2010- 08-12	15:43:00	F9 v1.0 B0004	CCAFS LC- 40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
	2012- 08-10	00:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
	2013- 01-03	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt
	2013- 03-12	22:41:00	F9 v1.1	CCAFS LC- 40	SES-8	3170	GTO	SES	Success	No attempt

#### Display 5 records where launch sites begin with the string 'CCA'

In [30]: %sql select \* from Spacex where launch\_site like 'CCA%' limit 5

\* ibm\_db\_sa://ybt98819:\*\*\*@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb Done.

Out[30]:	DATE	timeutc_	booster_version	launch_site	payload	payload_masskg_	orbit	customer	mission_outcome	landing_outcome
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	2012- 08-10	00:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
	2013- 01-03	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt
	2013- 03-12	22:41:00	F9 v1.1	CCAFS LC- 40	SES-8	3170	GTO	SES	Success	No attempt

Task 3

#### Display the total payload mass carried by boosters launched by NASA (CRS)

In [33]: %sql select sum(PAYLOAD\_MASS\_\_KG\_) from spacex where customer = 'NASA (CRS)'

 $* ibm\_db\_sa://ybt98819:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludbDone.$ 

Out[33]: 1

'

22007

#### Task 4

Display average payload mass carried by booster version F9 v1.1

#### Task 5

List the date when the first successful landing outcome in ground pad was acheived.

Hint:Use min function

#### 

# Task 7 List the total number of successful and failure mission outcomes In [47]: %sql select count(mission\_outcome) from Spacex where mission\_outcome like 'Success%' \* ibm\_db\_sa://ybt98819:\*\*\*@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb Done. Out[47]: 1 45 In [54]: %sql select count(mission\_outcome) from Spacex where mission\_outcome like 'Failure%' \* ibm\_db\_sa://ybt98819:\*\*\*@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb Done. Out[54]: 1 0

#### Task 8

List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery

```
9]: %sql select booster_version from Spacex where payload_mass__kg_ = (select max(payload_mass__kg_) from Spacex)
   * ibm_db_sa://ybt98819:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.

9]: booster_version
   F9 B5 B1048.4
   F9 B5 B1049.5
   F9 B5 B1060.2
   F9 B5 B1060.2
   F9 B5 B1058.3
```

#### Task 9

List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015

```
3]: %sql select launch_site, booster_version from Spacex where date like '2015-%' and landing__outcome like 'Failure (drone ship)'
    * ibm_db_sa://ybt98819:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb Done.

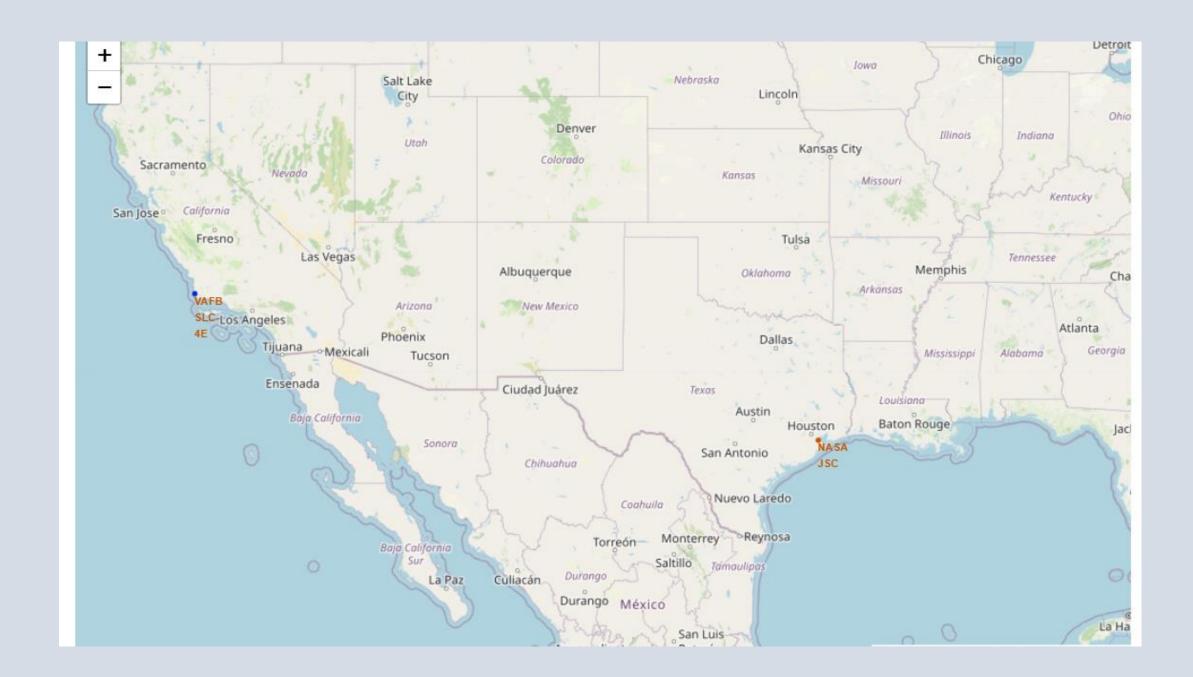
3]: launch_site booster_version
    CCAFS LC-40 F9 v1.1 B1012
```

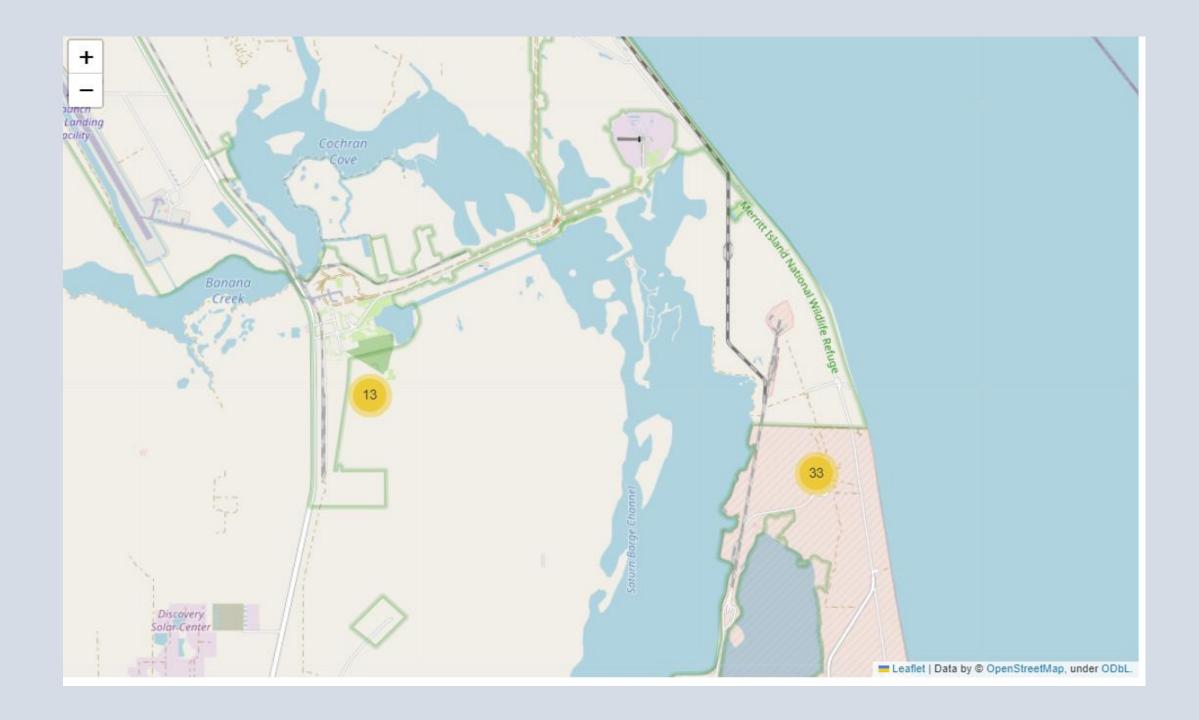
#### Task 10

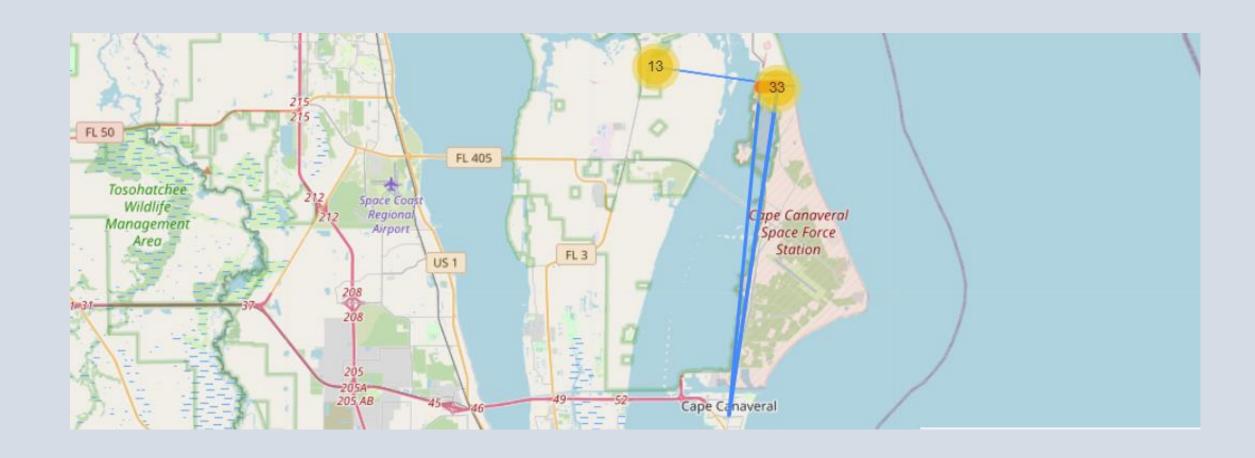
Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order



# Folium visualization



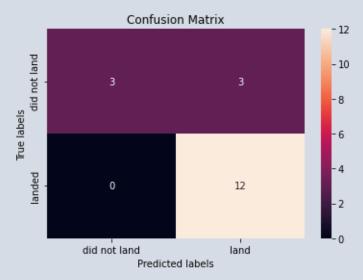




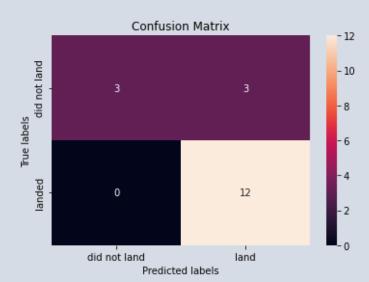
```
In [155]: # distance calc.
          coordinates3=[
              [28.56342, -80.57674],
              [28.39475, -80.60459]]
          distance_city = calculate_distance(coordinates3[0][0], coordinates3[0][1], coordinates3[1][0], coordinates3[1][1])
          distance_city
Out[155]: 18.957699612455393
In [156]: coordinates2 = [
                  [28.56414, -80.57097],
                  [28.56342, -80.57674]]
          distance_highway = calculate_distance(coordinates2[0][0], coordinates2[0][1], coordinates2[1][0], coordinates2[1][1])
          distance_highway
Out[156]: 0.5693408760028155
In [157]: coordinates1 = [
                  [28.56282, -80.58695],
                  [28.56342, -80.57674]]
          distance_railway = calculate_distance(coordinates1[0][0], coordinates1[0][1], coordinates1[1][0], coordinates1[1][1])
          distance_railway
Out[157]: 0.9996670765631016
```

# Machine Learning Prediction

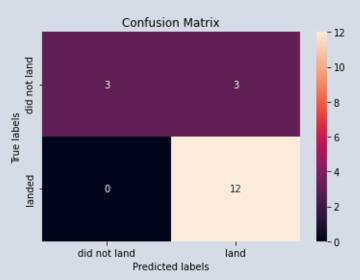
#### Log. Regression.



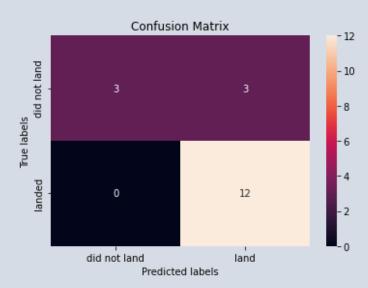
#### D. Tree



#### SVM



#### KNN



#### Log. Regression.

```
tuned hpyerparameters :(best parameters) {'C': 0.01, 'penalty': 'l2', 'solver': 'lbfgs'} accuracy : 0.8464285714285713
```

#### **SVM**

```
tuned hpyerparameters :(best parameters) {'C': 1.0, 'gamma': 0.03162277660168379, 'kernel': 'sigmoid'} accuracy : 0.8482142857142856
```

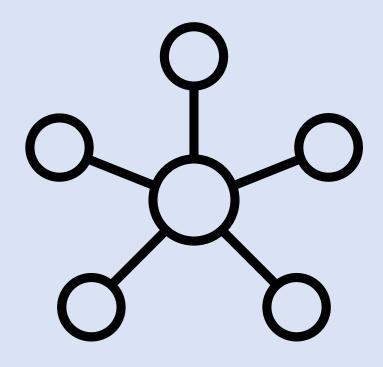
#### D. Tree

```
tuned hpyerparameters :(best parameters) {'criterion': 'entropy', 'max_depth': 4, 'max_features': 'sqrt', 'min_samples_leaf':
4, 'min_samples_split': 5, 'splitter': 'random'}
accuracy : 0.8892857142857142
```

#### KNN

```
tuned hpyerparameters :(best parameters) {'algorithm': 'auto', 'n_neighbors': 10, 'p': 1} accuracy : 0.8482142857142858
```

The best method has the value of: 0.8892857142857145
The method is Tree



## Conclusion

The biggest number of failed missions is between 2000-6000 kg of pay load mass.

The launch-site with the highest number of successful flights is CCAFS40.

The most succeful missions were on these orbits (ES-LI, GEO, HEO, SSO). However there was not enough data to support it therefore the most successful missions were on ULEO orbit.

The most successful booster versions between 4000-6000 kg of pay load mass are F9 FT B1 022 and F9 FT 1031 2. The total number of successful missions is 45. There is also the same number of success and failure on ground pads.

The majority of launchsites are 18 km close to city, 1 m close to railway and 0.6 km close to highway.

The best method for the prediction is the Tree with accuracy of 0.89.

The results are discussed on the next slide.

#### Discussion

• Considering all the findings if the next mission will be on the launch site CCAFS40, on ULEO orbit and with more payload than 6000 kg it with any starter, it will most likely be successful and therefore it will land and stage could be reused. It will also be possible to use less than 6000 kg of payload mass, but the only possible starter versions will be FT B1 022 and F9 FT 1031 2. However, the stage is close to railroads, highways and cities which might make the success of landing lower. On the other hand, most of the missions landed and we could predict whether another stage will be successful or not by using Tree method. If these factors will be kept in mind in future the costs should be minimal and missions should be successful.



# Appendix

- Link to Github repository <u>LINK</u>
- Course LINK